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Citation for published version (APA):
Ezzeldin Mahdy Abdelmonem, M., Jokic, A., & Bosch, van den, P. P. J. (2010). A parameter varying lyapunov function approach for tracking control for Takagi-Sugeno class of nonlinear systems. In Proc. 8th IEEE Int. conference on Control & Automation (ICCA ‘10), 09-11 June 2010, Xiamen, China (pp. 428-433). Institute of Electrical and Electronics Engineers. https://doi.org/10.1109/ICCA.2010.5524409

DOI:
10.1109/ICCA.2010.5524409

Document status and date:
Published: 01/01/2010

Document Version:
Publisher’s PDF, also known as Version of Record (includes final page, issue and volume numbers)

Please check the document version of this publication:
• A submitted manuscript is the version of the article upon submission and before peer-review. There can be important differences between the submitted version and the official published version of record. People interested in the research are advised to contact the author for the final version of the publication, or visit the DOI to the publisher’s website.
• The final author version and the galley proof are versions of the publication after peer review.
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Link to publication
A Parameter Varying Lyapunov Function Approach for Tracking Control for Takagi-Sugeno Class of Nonlinear Systems

M. Ezzeldin, A. Jokic and P.P.J. van den Bosch

Abstract— A reference model tracking control technique for nonlinear systems based on the Takagi-Sugeno model is proposed. The control design synthesis is aimed to reduce the tracking error for all bounded reference inputs and disturbances and to guarantee $\mathcal{L}_2$ gain performance. A nonlinear static output feedback controller is proposed to tackle this problem. Unlike the approaches using a single quadratic Lyapunov function, a parameter varying quadratic Lyapunov function is employed in our approach. The controller synthesis is formulated in terms of a feasibility problem of a set of linear matrix inequalities, which can be efficiently solved. A simulation example of a two-link robot system demonstrates the tracking performance and the validity of the proposed approach.

I. INTRODUCTION

The Takagi-Sugeno (T-S) model-based approach has nowadays become popular since it showed its efficiency to control complex nonlinear systems and has been used for many applications. The Takagi-Sugeno (T-S) models were introduced in [1], with the purpose of approximating complex nonlinear systems and can be described as the weighted sum of local linear models. The weights depend on the working point of the nonlinear system. The T-S models can be seen as a generalization of piece-wise linear systems. Based on the so called Universal Approximation theorem [2], appropriate weighting of the local models can reproduce the original nonlinear system arbitrary well. T-S models have led to the design of a feedback controller for each local model, which are further combined into a single overall controller. Stability and stabilization analyses, of the T-S model, have been investigated through Lyapunov direct method [3]-[5]. The key point of the proposed approaches is to achieve conditions in terms of LMIs (linear matrix inequalities). For different approaches on this subject the interested reader is referred to [6]-[10].

Tracking control design is an important issue for practical applications, for example, in robotic tracking control, missile tracking control and attitude tracking control of aircraft. However, there are very few studies concerning tracking control design based on the T-S model [11]-[14]. Most of the available tracking control techniques using T-S models are either based on state or observer based output feedback and based on the conventional, single quadratic Lyapunov function. These techniques depend on finding a constant positive definite matrix of a quadratic Lyapunov function to satisfy the stability condition of the system. In [14] a fuzzy tracking controller design for discrete time systems is proposed using the feedback linearization technique. Similarly, [15] has established a simple necessary and sufficient condition to determine local stability for the fuzzy systems and has derived a fuzzy tracking controller. In [11], an $H_\infty$ performance, related to the tracking error for bounded reference inputs, is formulated and an observer-based fuzzy controller is developed.

In this paper, we consider the tracking problem for nonlinear systems described by a T-S model with external disturbances. A reference model tracking based static output feedback controller is proposed to achieve minimal tracking error in terms of the desired $\mathcal{L}_2$ gain tracking performance. Instead of using the conventional quadratic Lyapunov function, in our approach a parameter varying quadratic Lyapunov function is employed. The parameter varying Lyapunov function approach for stabilization problem was already studied in [7], where the considered Lyapunov function is a weighted sum of several quadratic Lyapunov functions. The weights are the same ones as used in the T-S model. Note that this Lyapunov function, which is used in our approach as well, presents a significantly wider class of functions than the conventional quadratic one. Initial formulation of the considered $\mathcal{L}_2$ gain tracking problem is necessarily given in terms of a nonlinear matrix inequality. A transformation is proposed to transform this nonlinear inequality into a set of LMIs for which efficient optimization techniques are available. Furthermore, the stability of the proposed controller is guaranteed for a general class of T-S systems.

The main contributions of this paper are summarized as follows. The $\mathcal{L}_2$ gain static output feedback control design for the model tracking control is extended from linear system toward nonlinear system using nonlinear control. Moreover, based on the proposed approach, a simple algorithm in terms of LMI optimization technique is developed.

The remainder of the paper is organized as follows. Section II presents basic definitions of the T-S model and the problem formulation, while Section III introduces $\mathcal{L}_2$ gain based tracking control. Section IV presents the LMI formulation of the proposed approach. A simulation example is given in...
section V to illustrate the design effectiveness. Concluding remarks are collected in section VI.

II. PROBLEM FORMULATION

In this section, we recall the definition and some properties of the T-S model which is used to approximate a class of nonlinear system in the form:

\[
\dot{x}(t) = f(x(t)) + g(x(t))u(t) + B_w w(t), \\
y(t) = Cx(t),
\]

(1)

where \( x(t) \in \mathbb{R}^n \) denotes the state vector, \( u(t) \in \mathbb{R}^m \) denotes the control input, \( w(t) \in \mathbb{R}^{m_w} \) denotes the unknown but bounded disturbance, the vector fields \( f(x(t)) \) and \( g(x(t)) \) are smooth functions with \( f(0) = 0 \), and \( y(t) \in \mathbb{R}^p \) is the output of the system which depends linearly on the states with \( C \in \mathbb{R}^{p \times n} \).

Next, we will formally present the T-S model as a weighed average of several linear models through weighting functions. First, consider a set of linear models of the form:

\[
\dot{x}(t) = A_i x(t) + B_i u(t) + B_{iw} w(t), \\
y(t) = C_i x(t),
\]

(2)

for \( i = 1, \cdots, L \), where \( A_i \in \mathbb{R}^{n \times n} \), \( B_i \in \mathbb{R}^{n \times m} \), \( C_i \in \mathbb{R}^{p \times n} \), and \( L \) is the number of linear models. The T-S model is defined as:

\[
\Sigma : \begin{align*}
\dot{x}(t) &= \sum_{i=1}^{L} h_i(z(t)) \left( A_i x(t) + B_i u(t) + B_{iw} w(t) \right), \\
y(t) &= \sum_{i=1}^{L} h_i(z(t)) C_i x(t),
\end{align*}
\]

(3)

with \( h_i(z(t)) \geq 0 \) for \( i = 1, \cdots, L \), \( \sum_{i=1}^{L} h_i(z(t)) = 1 \) and \( z(t) \) are triggering variables, which depend on the current value of system states \( x(t) \). In this paper, we will assume that \( C_i = C \) for \( i = 1, \cdots, L \), what is in conformance with the output of the original nonlinear system (1).

**Remark 1:** An example for the membership functions \( h_i(z(t)) \) is shown in Fig. 1. Since in this paper we are concerned with the output feedback control, the triggering variables \( z(t) \) are considered to be dependent only on the measured output \( y(t) \).

The T-S model (3) is a general nonlinear time-varying dynamic system. Without going into more details, the T-S model (3) can capture the piecewise linear system (PWL) accurately if \( h_i(z(t)) \) belongs to the discrete set \( \{0, 1\} \).

To define the output tracking problem, we first introduce a linear time invariant reference model given by:

\[
\begin{align*}
x_r &= A_r x_r(t) + B_r r(t), \\
y_r &= C_r x_r(t),
\end{align*}
\]

(4)

where \( x_r(t) \) is the reference state, \( A_r \) is an asymptotically stable matrix, \( r(t) \) is a bounded reference input and \( y_r(t) \) is the reference output. It is assumed that, for all \( t \geq 0 \), \( y_r(t) \) represents a desired trajectory for \( y(t) \) to follow. The choice of the reference model parameters mainly depends on the desired closed loop behavior, e.g. system overshoot, settling time, etc. For more details about how to choose the reference model, see [16].

III. OPTIMIZED \( \mathcal{L}_2 \) GAIN BASED TRACKING CONTROL

In this paper, we are concerned with the controller synthesis which incorporates the reference model tracking performance criteria. Next, we recall the definition of dissipativity and the dissipativity characterization of the upper bounds for the \( \mathcal{L}_2 \) gain.

**Definition 1:** The system \( \Sigma \) with state \( x \in \mathbb{R}^n \), input \( w \in \mathbb{R}^{m_w} \) and output \( y \in \mathbb{R}^p \), is dissipative with respect to the supply function \( s : \mathbb{R}^{m_w} \times \mathbb{R}^p \to \mathbb{R} \) if there exists a positive definite continuous function \( V : \mathbb{R}^n \to \mathbb{R}_+ \), called a storage function, such that \( V(x(t_1)) - V(x(t_0)) \leq \int_{t_0}^{t_1} s(w(t), y(t)) dt \) for all \( t_1 \geq t_0 \), and for all corresponding solutions \( (w, x, y) \) to the system \( \Sigma \) on the interval \([t_0, t_1]\).

**Proposition 1:** The system \( \Sigma \) has a finite \( \mathcal{L}_2 \)-gain from input \( w \) to the output \( y \) smaller than or equal to \( \rho \) if it is dissipative with respect to the supply function \( s(w, y) = \rho^2 w^\top w - y^\top y \).

Next we present the structure of the output feedback controller considered in this paper. Consider a set of static feedback controllers of the following form:

\[
u(t) = K_j [y(t) - y_r(t)] \quad \text{for} \quad j = 1, \cdots, L.
\]

Then the overall controller is defined by:

\[
u(t) = \sum_{j=1}^{L} h_j(z(t)) K_j [y(t) - y_r(t)],
\]

(5)

where \( K_j \) are the controller gains for \( j = 1, \cdots, L \) and the functions \( h_j(z(t)) \), for \( j = 1, \cdots, L \) are the same ones as in (3).

The closed loop system can be expressed as the following

![Normalized membership functions.](image)
augmented system:

\[
\begin{bmatrix}
\dot{x}(t) \\
\dot{x}_r(t)
\end{bmatrix} = L \sum_{i=1}^{L} L \sum_{j=1}^{L} h_i(z(t)) h_j(z(t)) \\
\begin{bmatrix}
A_i + B_i K_j C - B_i K_j C_r \\
0
\end{bmatrix} \\
\begin{bmatrix}
x(t) \\
x_r(t)
\end{bmatrix} + \begin{bmatrix}
B_{\alpha} & 0 \\
0 & B_r
\end{bmatrix} \begin{bmatrix}
w(t) \\
r(t)
\end{bmatrix}
\]

Proof: Set \( \tilde{w} = 0 \) and consider the candidate Lyapunov function

\[
V(\tilde{x}(t)) = L \sum_{k=1}^{L} h_k \tilde{x}_r(t)^T P_k \tilde{x}(t).
\]

Furthermore, using (7) we have

\[
\dot{V}(\tilde{x}(t)) = L \sum_{k=1}^{L} L \sum_{i=1}^{L} h_i h_k (\tilde{x}_r(t) \tilde{A}_{ij} P_k \tilde{x}(t) + \tilde{x}_r(t) P_k \tilde{A}_{ij} \tilde{x}(t)) \\
+ \tilde{w}(t) E_{ij}^T P_k \tilde{x}(t) + \tilde{x}_r(t) P_k E_{ij} \tilde{w}(t) + \frac{1}{\rho^2} P_k E_{ij}^T P_k \tilde{x}(t) + \rho^2 \tilde{w}_r(t) \tilde{w}(t).
\]

Theorem 1: Suppose that \( |h_k| \leq \lambda_k \), \( \lambda_k > 0 \). Given the closed loop system (7), if there exist symmetric positive definite matrices \( P_k \) as a solution of the following matrix inequalities:

\[
\tilde{A}_{ij}^T P_k + P_k \tilde{A}_{ij} + \tilde{P}_k + \frac{1}{\rho^2} P_k E_{ij}^T P_k + \tilde{Q} < 0, \quad i, j, k = 1, \ldots, L
\]

with \( \tilde{P}_k = \sum_{k=1}^{L} \lambda_k (P_k + Y) \), and where \( \lambda_k \) are scalars, \( Y \) is a symmetric matrix, then the \( \mathcal{L}_2 \) gain tracking performance in (8) is guaranteed for a prescribed value of \( \rho \).

Proof: Consider the candidate Lyapunov function:

\[
\dot{V}(\tilde{x}(t)) = \sum_{k=1}^{L} \frac{1}{\lambda_k} h_k (\tilde{x}(t))^T P_k \tilde{x}(t),
\]

where the function \( h_k(z(t)) \), for \( k = 1, \ldots, L \), are the same ones as in (3).

The time derivative of \( V(t) \) along the systems trajectories is given by:

\[
\dot{V}(\tilde{x}(t)) = \frac{1}{\lambda_k} \sum_{k=1}^{L} h_k (\tilde{x}(t))^T P_k \tilde{x}(t) + \sum_{k=1}^{L} \frac{1}{\lambda_k} h_k (\tilde{x}(t))^T P_k \tilde{x}(t) \\
+ \sum_{k=1}^{L} \frac{1}{\lambda_k} h_k (\tilde{x}(t))^T P_k \tilde{x}(t).
\]
Then we can write the following equality
\[
\dot{V}(\tilde{x}(t)) = \sum_{k=1}^{L} h_k (\dot{\tilde{x}}^\top(t)P_k \tilde{x}(t) + \tilde{x}^\top(t)P_k \dot{\tilde{x}}(t) + \dot{h}_k)
\]
Using (9), from the above inequality we deduce that
\[
\dot{V}(\tilde{x}(t)) < 0
\]
In the next section, we will present a constructive algorithm to compute the controller gains based on Theorem 1 in terms of an LMI feasibility problem.

IV. LMI FORMULATION

A class of numerical optimization problems, called linear matrix inequality (LMI) problems, has received significant attention over the last decade [17]-[19]. These optimization problems can be solved in polynomial time via the interior-point methods [18], and hence are tractable, at least in a theoretical sense.

Note that (9) is not an LMI in \( P_k \) and \( K_i \) as variables. Therefore, a transformation method is proposed to reformulate (9) as a feasibility problem of a set of LMIs. This transformation method is summarized in the following Lemma.

**Lemma 1:** The following LMI implies the inequality (9):
\[
\begin{bmatrix}
-P_k \bar{Q} - \bar{Q}^\top P_k + \bar{P}_k + \bar{Q} \\
\frac{1}{2P_k E_i E_i^\top} P_k + A_{ij} + \bar{Q} + \frac{1}{\rho^2} P_k - \frac{2}{\rho^2} I
\end{bmatrix} < 0, \quad (15)
\]
for \( i, j, k = 1, \ldots, L \).

**Proof:** Pre- and Post-multiplying (15) by \([I \ P_k]\) and \([I \ -P_k]\) respectively, yields
\[
\tilde{A}_{ij} P_k + P_k \tilde{A}_{ij} + \bar{P}_k + \bar{Q} + \frac{1}{\rho^2} P_k E_i E_i^\top P_k + \bar{Q} < 0
\]
Note that (15) is an LMI in \( P_k \) and \( K_i \) and can therefore be efficiently solved by a dedicated software.

According to the above analysis, the tracking control via nonlinear static output feedback is now summarized as follows:

**Design Procedure:**
1) Select weighting functions, \( h_i(z(t)) i = 1, 2, \ldots, L \), and construct a T-S model as in (3).
2) Select an initial value of \( \rho^2 \) and \( \lambda_k \).
3) Solve the LMI problem in (15).
4) Decrease \( \rho^2 \) and repeat steps 3-4 until the LMI problem can not be solved.
5) Construct the controller (5) and verify whether \( |h_k| \leq \lambda_k \).

**Remark 2:** The time derivative of (10) contains information on the time derivative of the weighting functions \( h_k \). It’s assumed that the time derivative of the weighting functions \( h_k \) are upper bounded with a given value, which may lead to some conservatism in the LMI problem. However, the use of structural information on T-S model to construct the Lyapunov function is an advantage in comparison with disregarding the time derivative of the weighting functions. Moreover, the stability proof is based on this assumption, hence, it should be checked that this assumption is satisfied to have a complete stability proof. Furthermore, note that the number of LMIs (15) are \( L^3 \) which will have a high computational time, however, the number of the LMIs can be further reduced to \( L^3 - 2L^2 + 2L \) taking into account that not all the T-S subsystems are triggered at the same time [5].

V. SIMULATION EXAMPLE

To illustrate the proposed design approach and show its efficiency, the results are applied to the two-link robot system. The dynamic equations of the two-link robot system are given as follows [11]:
\[
M(q) \ddot{q} + C(q, \dot{q}) \dot{q} + G(q) = T
\]
where
\[
M(q) = \begin{bmatrix}
(m_1 + m_2) l_1^2 & m_2 l_1 l_2 (s_1 s_2 + c_1 c_2) \\
m_2 l_1 l_2 (c_1 s_2 - s_1 c_2) & m_2 l_2^2
\end{bmatrix}
\]
\[
C(q, \dot{q}) = \begin{bmatrix}
0 & -\dot{q}_2 \\
-\dot{q}_1 & 0
\end{bmatrix}
\]
\[
G(q) = \begin{bmatrix}
-(m_1 + m_2) l_1 g s_4 \\
-m_2 l_2 g s_2
\end{bmatrix}
\]
and the system output \( y = q = [q_1, q_2]^\top \), \( q_1, q_2 \) are generalized coordinates, \( M(q) \) is the moment of inertia, \( C(q, \dot{q}) \) includes Coriolis, centrifugal forces and \( G(q) \) is the gravitational force. Other quantities are: link mass \( m_1, m_2 \) [kg], link length \( l_1, l_2 \) [m], angular position \( q_1, q_2 \) [rad], applied torques \( T = [T_1, T_2]^\top \) [Nm], the gravity \( g = 9.8 \) [m/s²] and \( s_1 = \sin(q_1), s_2 = \sin(q_2), c_1 = \cos(q_1), c_2 = \cos(q_2) \).

Let \( x_1 = q_1, x_2 = \dot{q}_1, x_3 = \dot{q}_2 \) and \( x_4 = \ddot{q}_2 \), then the Takagi-Sugeno model which approximates the nonlinear system (16) can be represented using the following set of linearized models [11]:

<table>
<thead>
<tr>
<th>Case</th>
<th>( -\frac{\pi}{2} \leq x_1 \leq \frac{\pi}{2} ) and ( -\frac{\pi}{2} \leq x_3 \leq \frac{\pi}{2} )</th>
</tr>
</thead>
<tbody>
<tr>
<td>R1</td>
<td>( \ddot{x} = A_1 x + B_1 u + w, z = C_1 x )</td>
</tr>
<tr>
<td>R2</td>
<td>( -\frac{\pi}{2} \leq x_1 \leq \frac{\pi}{2} ) and ( -\frac{\pi}{2} \leq x_3 \leq \frac{\pi}{2} )</td>
</tr>
<tr>
<td>R3</td>
<td>( \ddot{x} = A_2 x + B_2 u + w, z = C_2 x )</td>
</tr>
<tr>
<td>R4</td>
<td>( -\frac{\pi}{2} \leq x_1 \leq \frac{\pi}{2} ) and ( -\frac{\pi}{2} \leq x_3 \leq \frac{\pi}{2} )</td>
</tr>
<tr>
<td>R5</td>
<td>( \ddot{x} = A_3 x + B_3 u + w, z = C_3 x )</td>
</tr>
<tr>
<td>R6</td>
<td>( -\frac{\pi}{2} \leq x_1 \leq \frac{\pi}{2} ) and ( -\frac{\pi}{2} \leq x_3 \leq \frac{\pi}{2} )</td>
</tr>
<tr>
<td></td>
<td>( \ddot{x} = A_6 x + B_6 u + w, z = C_6 x )</td>
</tr>
</tbody>
</table>
R7 If $0 \leq x_1 \leq \frac{\pi}{2}$ and $-\frac{\pi}{2} \leq x_3 \leq 0$
then $\dot{x} = A_7 x + B_7 u + w, z = C_7 x$

R8 If $0 \leq x_1 \leq \frac{\pi}{2}$ and $-\frac{\pi}{2} \leq x_3 \leq 0$
then $\dot{x} = A_8 x + B_8 u + w, z = C_8 x$

R9 If $0 \leq x_1 \leq \frac{\pi}{2}$ and $0 \leq x_3 \leq \frac{\pi}{2}$
then $\dot{x} = A_9 x + B_9 u + w, z = C_9 x$

where,

$$x = [x_1 \ x_2 \ x_3 \ x_4]^T, \ u = [T_1 \ T_2]^T \text{ and}$$

$$w = [w_1 \ w_2 \ w_3 \ w_4]^T.$$

$$A_1 = \begin{bmatrix}
5.927 & -0.001 & -0.315 & -8.4 \times 10^{-6} \\
0 & 0 & 0 & 1 \\
-6.859 & 0.002 & 3.155 & 6.2 \times 10^{-6}
\end{bmatrix}$$
$$A_2 = \begin{bmatrix}
3.0428 & -0.0011 & 0.1791 & -0.0002 \\
0 & 0 & 0 & 1 \\
3.5436 & 0.0313 & 2.5611 & 1.14 \times 10^{-5}
\end{bmatrix}$$
$$A_3 = \begin{bmatrix}
6.2728 & 0.003 & 0.4339 & -0.0001 \\
0 & 0 & 0 & 1 \\
9.1040 & 0.0158 & -1.0574 & -3.2 \times 10^{-5}
\end{bmatrix}$$
$$A_4 = \begin{bmatrix}
6.4535 & 0.0017 & 1.2427 & 0.0002 \\
0 & 0 & 0 & 1 \\
-3.1873 & -0.306 & 5.1911 & -1.8 \times 10^{-5}
\end{bmatrix}$$
$$A_5 = \begin{bmatrix}
11.1336 & 0 & -1.8145 & 0 \\
0 & 0 & 0 & 1 \\
-9.0918 & 0 & 9.1638 & 0
\end{bmatrix}$$
$$A_6 = \begin{bmatrix}
6.1702 & -0.001 & 1.687 & -0.0002 \\
0 & 0 & 0 & 1 \\
-2.3559 & 0.0314 & 4.5298 & 1.1 \times 10^{-5}
\end{bmatrix}$$
$$A_7 = \begin{bmatrix}
6.1702 & -0.0041 & 0.6205 & 0.0001 \\
0 & 0 & 0 & 1 \\
8.8794 & -0.0193 & -1.0119 & 4.4 \times 10^{-5}
\end{bmatrix}$$
$$A_8 = \begin{bmatrix}
3.6421 & 0.0018 & 0.0721 & 0.0002 \\
0 & 0 & 0 & 1 \\
2.429 & -0.0305 & 2.9832 & -1.9 \times 10^{-5}
\end{bmatrix}$$
$$A_9 = \begin{bmatrix}
0 & 0 & 0 & 0 \\
3.2933 & -0.0009 & -0.2188 & -1.2 \times 10^{-5} \\
0 & 0 & 0 & 1 \\
-7.4649 & 0.0024 & 3.2693 & 9.2 \times 10^{-6}
\end{bmatrix}$$

$$B_4 = \begin{bmatrix}
0 & 0 & 0 & 0 \\
0.5 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0
\end{bmatrix}, \ B_5 = \begin{bmatrix}
0 & 0 & 0 & 0 \\
1 & -1 & 0 & 0 \\
0 & 0 & 0 & 0 \\
-1 & 2 & 0 & 0
\end{bmatrix}, \ B_6 = \begin{bmatrix}
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 \\
0.5 & 0 & 0 & 0 \\
0 & 0 & 0 & 0
\end{bmatrix}$$

$$B_7 = \begin{bmatrix}
0 & 0 & 0 & 0 \\
1 & 1 & 0 & 0 \\
0 & 0 & 0 & 0 \\
1 & 2 & 0 & 0
\end{bmatrix}, \ B_8 = \begin{bmatrix}
0 & 0 & 0 & 0 \\
1 & 1 & 0 & 0 \\
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0
\end{bmatrix}, \ B_9 = \begin{bmatrix}
0 & 0 & 0 & 0 \\
1 & 1 & 0 & 0 \\
0 & 0 & 0 & 0 \\
-1 & 2 & 0 & 0
\end{bmatrix}$$

$$C_i = \begin{bmatrix}
1 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 \\
0 & 0 & 0 & 1 \\
0 & 0 & 1 & 0
\end{bmatrix} \text{ for } i = 1, \ldots, 9$$

The triangle type of weighting functions for the triggering variables $x_1$ and $x_3$ in each rule are adopted for the T-S model as illustrated in Fig. 1 and $\lambda_k = 0.8$ for $k = 1, \ldots, L$. The reference model (4) parameters are given by [11]

$$A_r = \begin{bmatrix}
0 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 \\
0 & 0 & 0 & 1 \\
0 & 0 & -6 & -5
\end{bmatrix}, \ B_r = \begin{bmatrix}
1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 1 & 0 \\
0 & 0 & 0 & 1
\end{bmatrix}, \text{ and}$$

$$C_r = C_i.$$

Using Theorem 1 and the LMI optimization toolbox in Matlab, the following controller gains $K_j$ for $j = 1, \ldots, L$ are obtained

$$K_1 = \begin{bmatrix}
-602.5242 & -266.3152 \\
-266.3152 & -649.2914 \\
-547.3363 & -265.2308 \\
-265.2308 & -940.6244 \\
-486.4161 & -11.9060 \\
-11.9060 & -562.5966 \\
-670.0981 & -99.0489 \\
-99.0489 & -901.7503 \\
-782.7883 & -240.0344 \\
-240.0344 & -561.2339 \\
-634.0707 & -128.9711 \\
-128.9711 & -900.8147 \\
-438.1436 & 4.4070 \\
4.4070 & -570.1235 \\
-542.5231 & -224.0683 \\
-224.0683 & -945.0132 \\
-614.0204 & -254.7631 \\
-254.7631 & -645.2517
\end{bmatrix}$$

Different reference and disturbance signals are applied in simulation of the closed loop system. Fig. 2-3 present the simulation results for the proposed tracking control for T-S model. The simulation results indicate that the closed loop system can efficiently track the reference model in the presence of external disturbances and the desired performance can be achieved. It is found that the tracking error between the reference output $y_r(t)$ and actual closed loop output $y(t)$ is always less than 0.03 and $|\tilde{h}_k| \leq 0.4$ for $k = 1, \ldots, L$. 

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In this paper, the problem of reference model tracking of nonlinear systems has been considered. Based on the Takagi-Sugeno model, a nonlinear static output feedback controller is developed to reduce the tracking error by minimizing the attenuation level in terms of $\mathcal{L}_2$ gain. The novelty of the presented approach is that it is based on a parameter varying Lyapunov function. The advantage of the proposed tracking control design is that a simple static output feedback controller is used without feedback linearization technique and a complicated adaptive scheme. The complete synthesis problem has been formulated in terms of LMI feasibility which can be efficiently solved using a dedicated software. Simulation results based on a two-link robot arm have demonstrated the efficiency of the proposed algorithm.

VI. Conclusion

In this paper, the problem of reference model tracking of nonlinear systems has been considered. Based on the Takagi-Sugeno model, a nonlinear static output feedback controller is developed to reduce the tracking error by minimizing the attenuation level in terms of $\mathcal{L}_2$ gain. The novelty of the presented approach is that it is based on a parameter varying Lyapunov function. The advantage of the proposed tracking control design is that a simple static output feedback controller is used without feedback linearization technique and a complicated adaptive scheme. The complete synthesis problem has been formulated in terms of LMI feasibility which can be efficiently solved using a dedicated software. Simulation results based on a two-link robot arm have demonstrated the efficiency of the proposed algorithm.

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